

# Predicting Football Match Outcomes

Machine Learning Project - Bar Ilan University

- **Presented by:** Leonardo Romano
- **Course:** Machine Learning (Spring 2025)



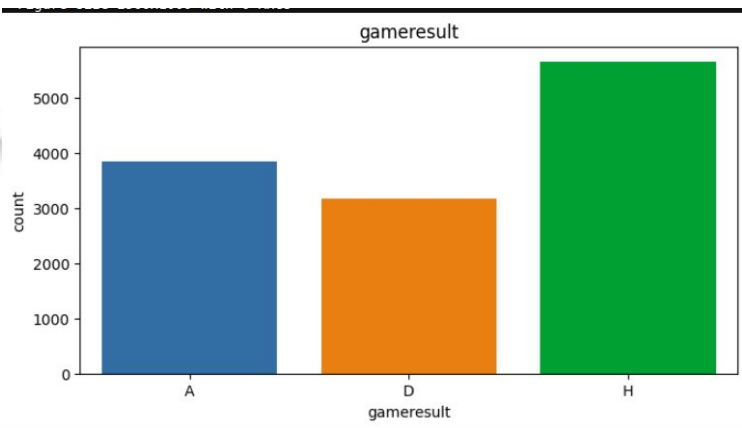
# Challenges

- Separated datasets
- Selecting the working environment
- Computer crashes
- Project workflow not so smooth



# Executive Summary

Objective: Predict match result  
(Home Win / Draw / Away Win)



- Dataset from Kaggle:  
<https://www.kaggle.com/datasets/technika148/football-database>
- Multi-step pipeline:
  - Data cleaning & feature engineering
  - Exploratory analysis & statistical testing
  - Model training & tuning
- Final model: XGBoost Classifier with  $>99\%$  accuracy

# Procedure

1. Linked PlayerID to TeamID
2. Date parsing & time feature extraction
3. EDA
4. Creating categorical features to handle outliers with low cardinality
5. Run MICE to fill missing values due to outliers removal
6. Feature Engineering and Selection
7. Model selection and final Model used XGboost



# Procedure

## 1. Linked PlayerID to TeamID

	playerID	teamID	playerName	teamName
0	560	89	Sergio Romero	Manchester United
1	557	89	Matteo Darmian	Manchester United
2	548	89	Daley Blind	Manchester United
3	628	89	Chris Smalling	Manchester United
4	1006	89	Luke Shaw	Manchester United
...	...	...	...	...
10101	7396	176	Loic Bessile	Bordeaux
10102	9566	175	Yanis Lhéry	Saint-Etienne
10103	9565	175	Mathys Saban	Saint-Etienne
10104	9568	181	Charles Costes	Dijon
10105	9567	181	Erwan Belhadji	Dijon

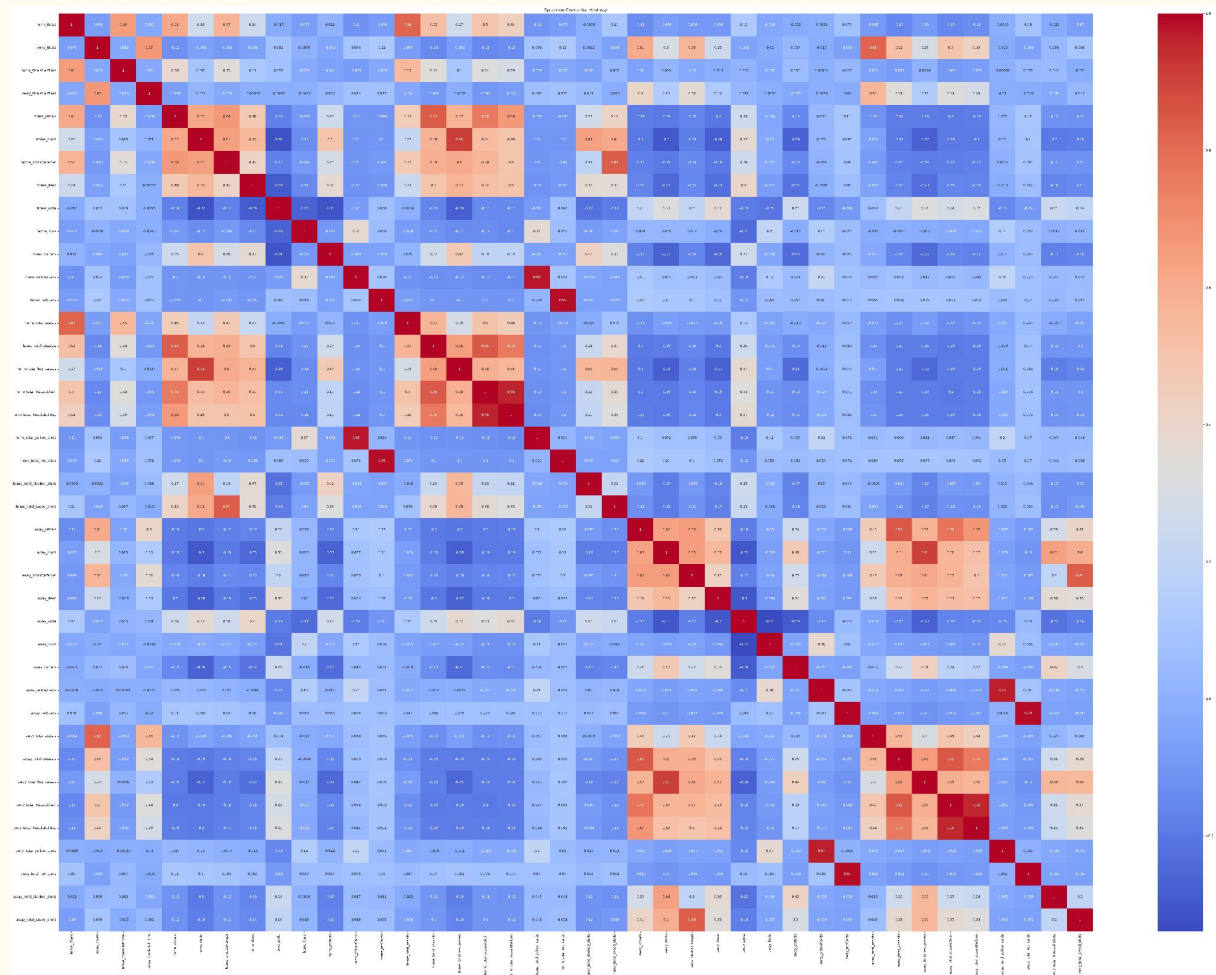
# Procedure

## 3. EDA-

- Class balance: Nearly even across 3 outcomes
- Key variables:
  - Goals, Shots, xG(goals and assist), Corners, Cards
- Statistical tests:
  - T-tests for numeric vars
  - Chi-square for categoricals

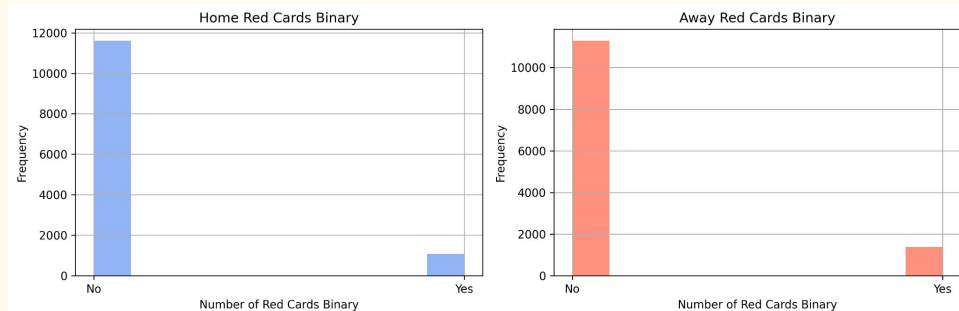
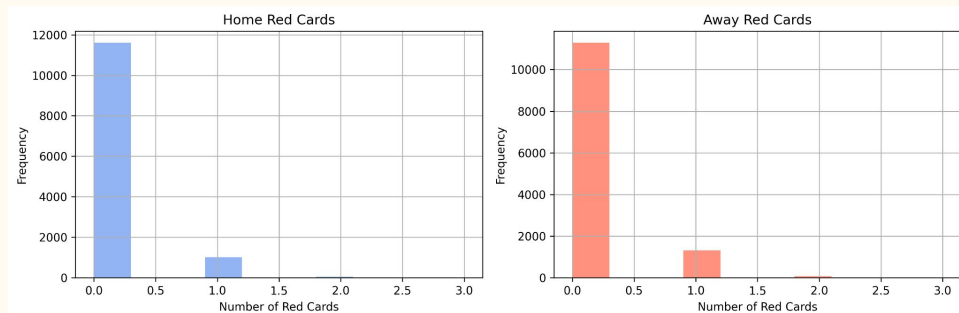
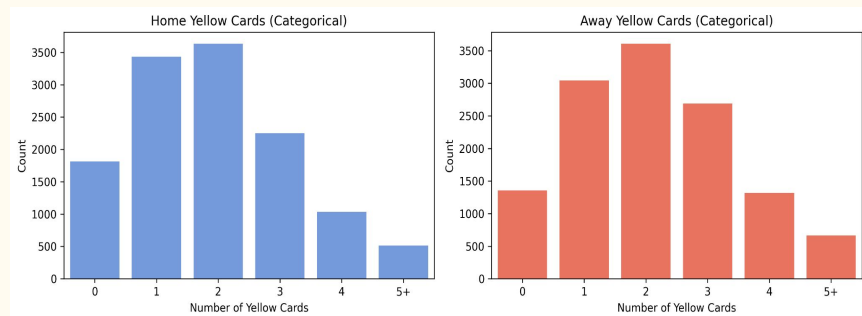
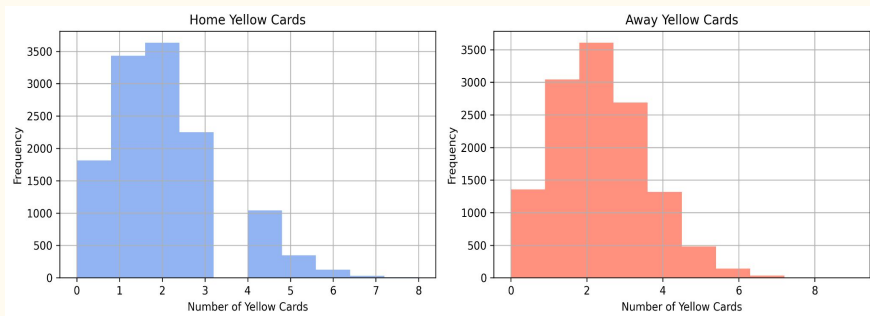


## Correlations

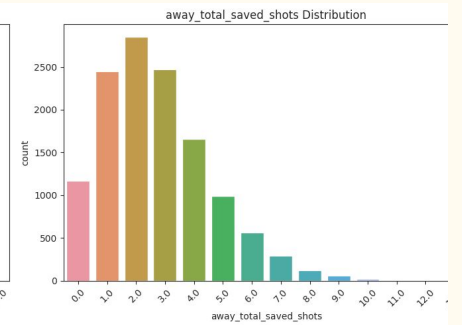
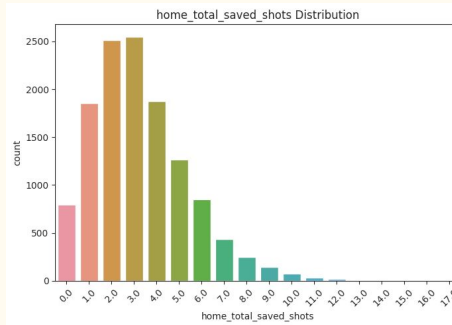
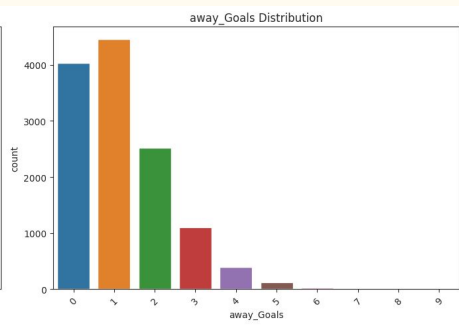
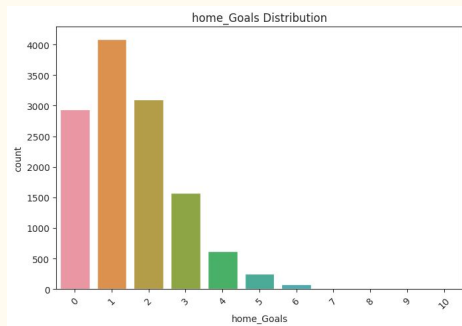
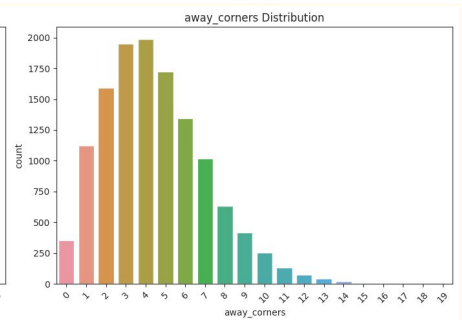
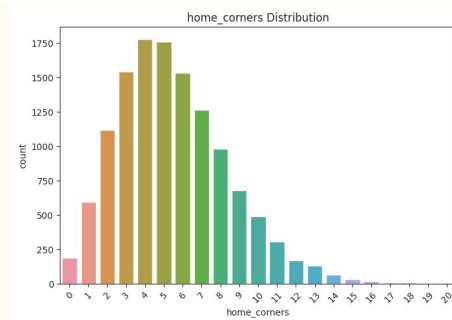
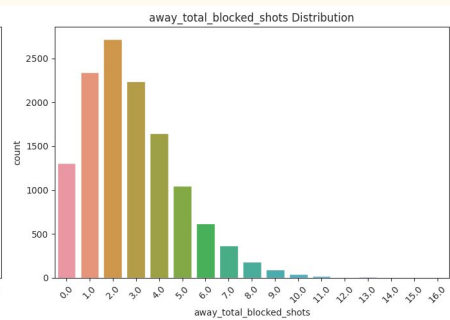
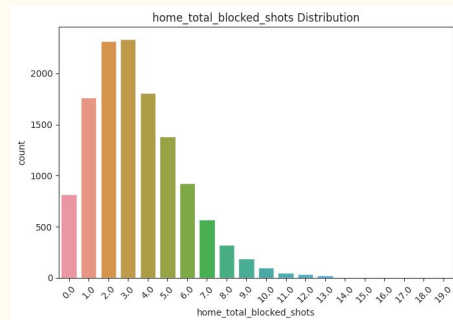
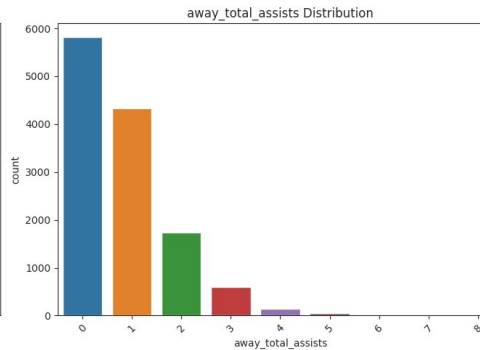
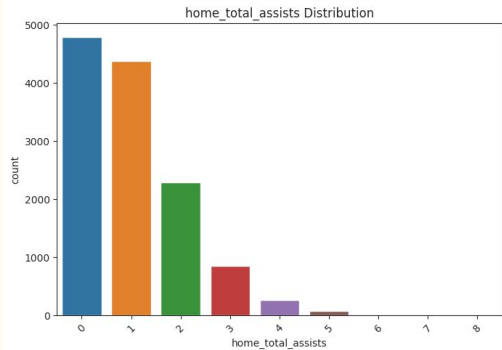
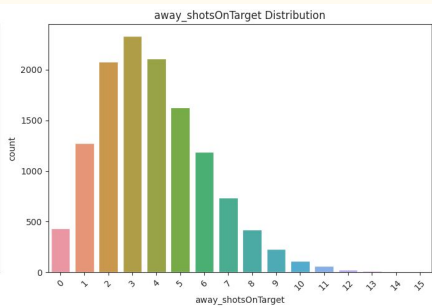
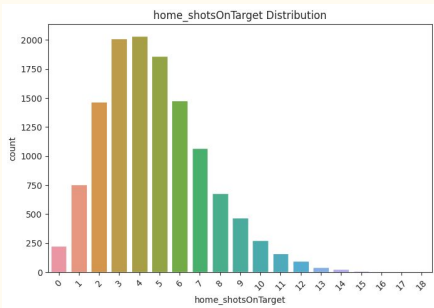


# Procedure

## 4. Creating categorical features to handle outliers with low cardinality

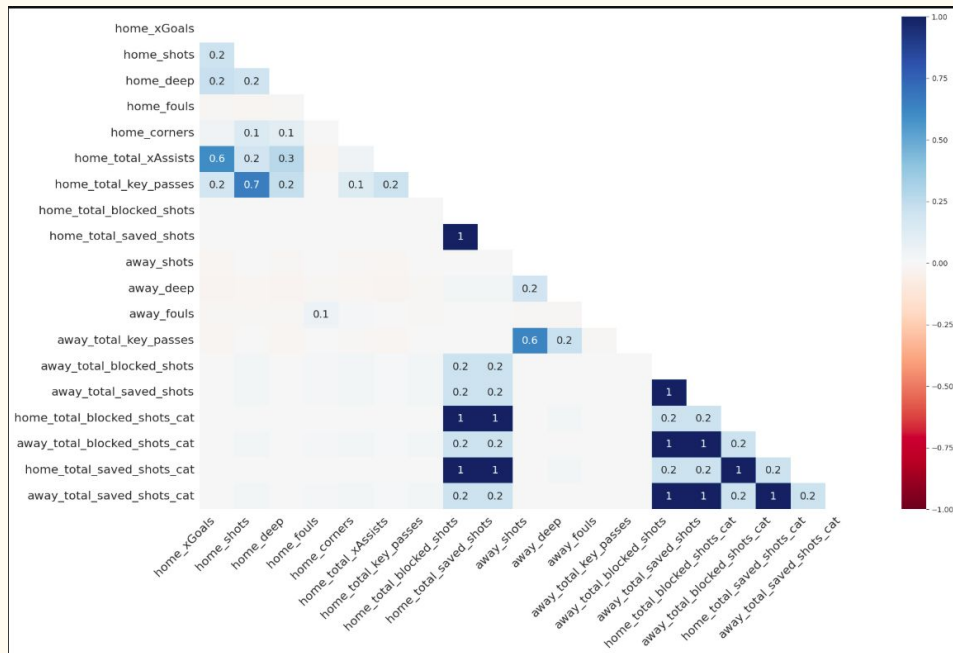






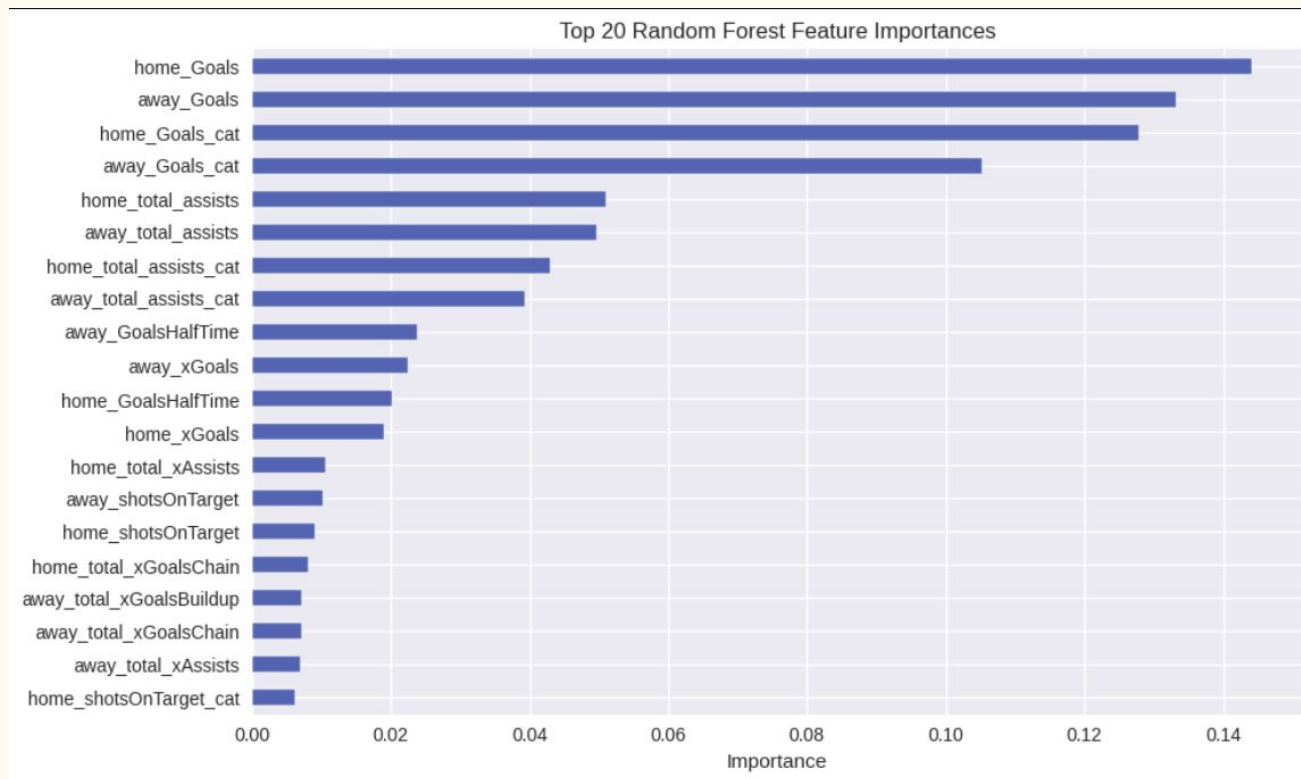
# Procedure

5. Run MICE to fill missing values due to outliers removal



## 6. Feature engineering

- Created 30+ features:
  - Rolling averages
  - Goal efficiency ratios
  - Disciplinary scores
- Plotted distributions of engineered features



## 6. Feature Selection

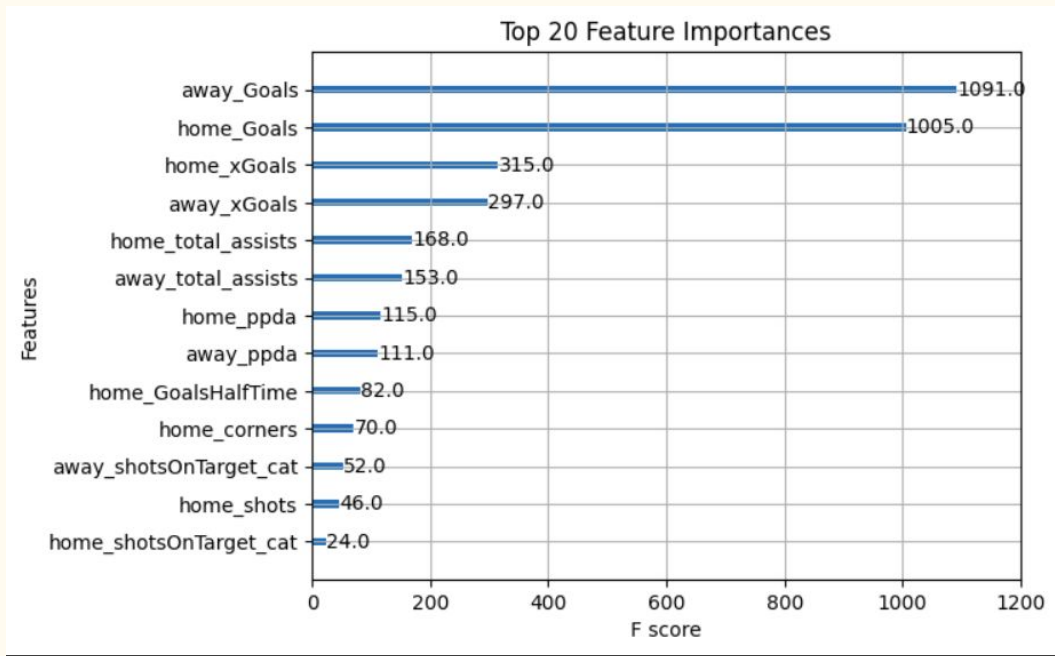
- Univariate tests (t-test, chi-squared)
- Multivariate: Lasso, Random Forest, Gradient Boosting
- Kept top 18 most important features

0	home_Goals	2536	non-null	int64
1	away_Goals	2536	non-null	int64
2	home_GoalsHalfTime	2536	non-null	int64
3	home_xGoals	2536	non-null	float64
4	home_shots	2536	non-null	float64
5	home_ppda	2536	non-null	float64
6	home_corners	2536	non-null	float64
7	home_total_assists	2536	non-null	int64
8	away_xGoals	2536	non-null	float64
9	away_ppda	2536	non-null	float64
10	away_total_assists	2536	non-null	int64
11	away_total_red_cards	2536	non-null	int64
12	home_shotsOnTarget_cat	2536	non-null	float64
13	away_shotsOnTarget_cat	2536	non-null	float64
14	home_total_assists_cat	2536	non-null	float64
15	away_total_assists_cat	2536	non-null	float64
16	home_Goals_cat	2536	non-null	float64
17	away_Goals_cat	2536	non-null	float64
18	gameresult	2536	non-null	int64
19	split	2536	non-null	object

## 7. Modeling and predictions

Model: **XGBClassifier**

	model	Accuracy	Precision	Recall	F1-score	Log-loss	AUC
	Logistic Regression	1.000000	1.000000	1.000000	1.000000	0.005162	1.000000
	XGB	0.998028	0.998034	0.998028	0.998026	0.005235	1.000000
	RandomForest	0.998028	0.998034	0.998028	0.998026	0.014983	0.999991
	GBM	0.999211	0.999213	0.999211	0.999211	0.007243	0.999971
	SVM	0.994479	0.994501	0.994479	0.994485	0.012243	0.999957
	Decision Tree	0.998028	0.998037	0.998028	0.998027	0.071064	0.998098
	ADABoost	0.662066	0.855655	0.662066	0.680578	0.645262	0.963105



The project is not complete - definitely  
there is still work to do

# Thanks

